TOWARDS BETTER PERCEPTION OF URBAN INFORMATION: A VISUALIZATION PERSPECTIVE.

by

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This is to certify that I have examined the above Ph.D. thesis and have found that it is complete and satisfactory in all respects, and that any and all revisions required by the thesis examination committee have been made.

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ABSTRACT

TBD

CHAPTER 1

INTRODUCTION

1.1 Motivation

The ongoing urbanization process has been one of the most important trend after the World War II (WWII). According to the recent urbanization report from United Nations [?], between 1950 and 2018, the estimated urban population increased more than fourfold, from 0.8 billion to 4.2 billion. The rapid urbanization also result in a series of problems such as the traffic congestion and environmental pollution, which attracts more and more attentions from the research field. Traditionally, it will take tremendous effort for researchers and urban planers to understand and explore the large scale urban dynamics due to the limited resource and existing cases.

Fortunately, the rapid development of information and communication technology(ICT) makes the urban data collection and analysis cheaper and easier, which provides the opportunities for people to study these problems and boost the development of relative discipline such as urban informatics [4] and urban computing [18]. For example, the street view images allow urban planners to conduct urban environment auditing without going to the field; the position data tracking and acquisition techniques embedded in the mobile devices allow researchers to capture the crowd movement pattern of urban residents; industrial emission and the meteorology data collected from the monitoring stations are able to help domain expert in precisely forecasting the air quality in future and provide the control strategy to the government.

Even though the computers have a great advantage in fast computing and large information storage, the fully automatics techniques still have limitations in analyzing the urban data due to the complexity and variety of the real world analysis tasks. It is always need the involvement of human beings who have acute perception and considerable experience in the domain field to make the final decision. Visualization, "the study of trans-

forming data, information, and knowledge into interactive visual representations" [9], bridges the gap between the human beings and the computing techniques.

In this chapter, we first introduce the existing visualization techniques in urban data exploration. Then we discuss the major contributions in this proposal which is followed by the proposal organization.

1.2 Visualization meets urban information

The analysis of model urban always refers to the intersection of three domains: people, place and technology [4], which is covered by the proposed work in this proposal.

- **People.** People refers to the residents, citizens, community groups as well as the relative data such as commuting, social network, etc.
- **Place.** Place refers to the physical environment such as the urban sites, locales and habitats which if formed by nature or human activities.
- Technology. Technology refers to a variety of techniques span informatics, computing techniques, wearable devices, etc.

Before introducing the contribution of this proposal, we first briefly introduce the existing techniques from these three aspects:

1.2.1 Visualization of human movement

As a general topic in visualization description, how to effectively visualizing the human movement has been long studied in recent years. An overview has been proposed to categorized the existing movement visualization techniques into three classes: **pattern extraction**, **direct depiction**, **summarization** [1,17].

The **pattern extraction** methods leverage effective visualization to discover the mobility pattern from the movement data such as the traffic congestions [14], route assessment [6,13], commuting patterns [2,11], co-occurrence [10,17]. Compare to other two types

of techniques, the pattern extract methods aim to solve the specific application problem directly and these techniques are always difficult to extend to other applications.

The **direct depiction** directly plot the movement data according to the data format: such as the line segments for OD trails, and polylines for trajectories [1,3,7]. Such techniques preserve all the information of individual movement. However, when the dataset is large, the visual clutter the render workload will be the major limitation which hinders the analysts' ability to explore the data.

The **summarization** techniques aim to provide an overview of movement data. Most of these summarization techniques conduct statistical calculation and aggregation before rendering. The typical visualization can be heatmap [15], density map [8], flow map [5], matrix [16] and node-link [12]. Due to the ability to present the high-level perception and alleviate the data uncertainty [1], the summarization techniques are widely used in movement visualization.

1.2.2 Visualization of urban environment

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1.2.3 Visual explanation for computing techniques

1.3 Contribution

In this proposal, we introduce several visualization techniques in assisting urban data exploration tasks as well as the understanding of urban computing techniques:

• Visual Exploration of Human-Scale Urban Forms. We propose StreetVizor, an interactive visual analytics system that helps planners leverage their domain knowledge in exploring human-scale urban forms based on street view images. Our system presents two-stage visual exploration: 1) an AOI Explorer for the visual comparison of spatial distributions and quantitative measurements in two areas-of-interest at city- and region-scales; 2) and a Street Explorer with a novel parallel coordinate

plot for the exploration of the fine-grained details of the urban forms at the streetscale. We integrate visualization techniques with machine learning models to facilitate the detection of street view patterns.

- An edge bundling technique for visualizing origin-destination trails in urban traffic. We propose RAEB(route aware edge bundling), a novel edge bundling technique to viusally summary the OD trails in urban traffic data. We identify inconsiderate settings of conventional kernel density estimation edge bundling (KDEEB) when applied to urban traffic data, including non-optimal kernel size and road neglect. The limitations are addressed in RAEB by introduction of a comprehensive pipeline comprising preprocessing, bundling and evaluation processes. A series of new parameters, together with adaptions of existing ones, are employed in the pipeline.
- Understanding recurrent neural networks on multi-dimensional time-series fore-cast. We propose MultiRNNExplorer, a visual analytics system to interpret RNNs on multi-dimensional time-series forecasts. Specifically, to provide an overview to reveal the model mechanism, we propose a technique to estimate the hidden unit response by measuring how different feature selections affect the hidden unit output distribution. We then cluster the hidden units and features based on the response embedding vectors. Finally, we propose a visual analytics system which allows users to visually explore the model behavior from the global and individual levels.

1.4 Proposal organization

The rest of this proposal is organized as follows:

Chapter 2 introduces the design of Streetvizor: specific analysis requirements are described by a collaborating urban planner, and the designs are evaluated and refined against requirements. We illustrate the applicability of our approach with case studies on the real-world datasets of four cities, i.e., Hong Kong, Singapore, Greater London and New York City. Interviews with domain experts demonstrate the effectiveness of our system in facilitating various analytical tasks.

Chapter 3 presents the details about the implementation of RAEB(route aware edge bundling), a new technique to visualize the origin-destination trails in traffic data by edge bundlings techniques. In the evaluation, we conducted experiments with artificial and real-world traffic data to demonstrate the advancements of RAEB in various applications, including the preservation of road network topology and support of multi-scale exploration.

Chapter 4 proposes a MultiRNNExplorer, a visual analytics system to interpret the recurrent neural networks on multi-dimensional time-series forecast. We introduce the system design and demonstrate the effectiveness of the proposed technique with case studies of air pollutant forecast applications.

Chapter 5 concludes the proposal and discusses possible future work.

CHAPTER 2 CONCLUSION AND FUTURE WORK

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